

# Example 7.4 Maximum likelihood estimation - Stochastic volatility models

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```
rm(list=ls(all=TRUE))
library(Quandl)

## Warning: package 'Quandl' was built under R version 3.2.2

## Loading required package: xts

## Warning: package 'xts' was built under R version 3.2.2

## Loading required package: zoo

## Warning: package 'zoo' was built under R version 3.2.2

##
## Attaching package: 'zoo'
##
## The following objects are masked from 'package:base':
##
##   as.Date, as.Date.numeric

library(DiffusionRgqd)
# Source data for the S&P500 index (SPX).
quandldata1 <- Quandl("YAHOO/INDEX_GSPC", collapse="weekly",
start_date="1990-01-01",end_date="2015-01-01", type="raw")
St <- rev(quandldata1[,names(quandldata1)=='Close'])
time1 <-rev(quandldata1[,names(quandldata1)=='Date'])
# Source data for the volatility index (VIX).
quandldata2 <- Quandl("YAHOO/INDEX_VIX", collapse="weekly",
start_date="1990-01-01",end_date="2015-01-01", type="raw")
Vt <- rev(quandldata2[,names(quandldata2)=='Close'])
time2 <- rev(quandldata2[,names(quandldata2)=='Date'])

GQD.remove() # Remove the previous model coefficients

## [1] "Removed : NA "

# R_t coefficients:
a00 <- function(t){theta[1]}
a01 <- function(t){-0.5*theta[2]*theta[2]}
c01 <- function(t){theta[2]*theta[2]}
d01 <- function(t){theta[2]*theta[5]*theta[6]}
# V_t coefficients:
```

```

b00 <- function(t){theta[3]}
b01 <- function(t){-theta[4]}
e01 <- function(t){theta[2]*theta[5]*theta[6]}
f01 <- function(t){theta[5]*theta[5]}

# Create data matrix and numerical time vector :
X <- cbind(log(St),(Vt/100)^2)
time <- cumsum(c(0,diff(as.Date(time1))*(1/365)))

# Some starting parameters for the optimization routine:
theta.start <- c(0,1,1,0.5,1,0)
# Calculate MLEs of the parameter vector:
model_1 <- BiGQD.mle(X,time,mesh=10,theta=theta.start)

```

## Compiling C++ code. Please wait.

```

## =====
##                               GENERALIZED QUADRATIC DIFFUSION
## =====
## ----- Drift Coefficients -----
## a00 : theta[1]
## a01 : -0.5*theta[2]*theta[2]
## ... ..
## b00 : theta[3]
## b01 : -theta[4]
## ----- Diffusion Coefficients -----
## c01 : theta[2]*theta[2]
## ... ..
## d01 : theta[2]*theta[5]*theta[6]
## ... ..
## e01 : theta[2]*theta[5]*theta[6]
## ... ..
## f01 : theta[5]*theta[5]
## ----- Model Info -----
## Time Homogeneous      : Yes
## Data Resolution       : Homogeneous: dt=0.0192
## # Removed Transits.   : None
## Density approx.       : 4th Ord. Truncation, Bivariate-Saddlepoint
## Elapsed time          : 00:00:13
## ... ..
## dim(theta)            : 6
## -----

```

```

# Retrieve parameter estimates and appr. 95% CIs:
GQD.estimate(model_1)

```

```

##           Estimate Lower_95 Upper_95
## theta[1]    0.083    0.031   0.135
## theta[2]    0.770    0.740   0.799
## theta[3]    0.168    0.128   0.207
## theta[4]    3.822    2.816   4.827

```

```
## theta[5]    0.431    0.414    0.447
## theta[6]   -0.671   -0.700   -0.641
```

```
GQD.remove() # Remove the previous model coefficients
```

```
## [1] "Removed :  a00 a01 b00 b01 c01 d01 e01 f01"
```

```
# R_t coefficients:
a00 <- function(t){theta[1]}
a02 <- function(t){-0.5*theta[2]*theta[2]}
c02 <- function(t){theta[2]*theta[2]}
d02 <- function(t){theta[2]*theta[5]*theta[6]}
# V_t coefficients:
b00 <- function(t){theta[3]}
b01 <- function(t){-theta[4]}
e02 <- function(t){theta[2]*theta[5]*theta[6]}
f02 <- function(t){theta[5]*theta[5]}

theta.start <- c(0,1,1,1,1,0)
model_2 <- BiGQD.mle(X,time,mesh=10,theta=theta.start)
```

```
## Compiling C++ code. Please wait.
```

```
## =====
##                      GENERALIZED QUADRATIC DIFFUSION
## =====
## ----- Drift Coefficients -----
## a00 : theta[1]
## a02 : -0.5*theta[2]*theta[2]
## ...   ...   ...   ...   ...   ...   ...   ...   ...   ...
## b00 : theta[3]
## b01 : -theta[4]
## ----- Diffusion Coefficients -----
## c02 : theta[2]*theta[2]
## ...   ...   ...   ...   ...   ...   ...   ...   ...   ...
## d02 : theta[2]*theta[5]*theta[6]
## ...   ...   ...   ...   ...   ...   ...   ...   ...   ...
## e02 : theta[2]*theta[5]*theta[6]
## ...   ...   ...   ...   ...   ...   ...   ...   ...   ...
## f02 : theta[5]*theta[5]
##
## ----- Model Info -----
## Time Homogeneous      : Yes
## Data Resolution       : Homogeneous: dt=0.0192
## # Removed Transits.   : None
## Density approx.       : 4th Ord. Truncation, Bivariate-Saddlepoint
## Elapsed time          : 00:00:40
## ...   ...   ...   ...   ...   ...   ...   ...   ...   ...
## dim(theta)            : 6
## -----
```

```
# Compare AIC and BIC vlaues for models 1 and 2:
GQD.aic(list(model_1,model_2))
```

```
##           Convergence p min.likelihood           AIC           BIC           N
## Model 1           0 6      -7852.212      -15692.424      -15661.381 1305
## Model 2           0 6      -7965.957  [=] -15919.914  [=] -15888.871 1305
```

```
GQD.remove()
```

```
## [1] "Removed :  a00 a02 b00 b01 c02 d02 e02 f02"
```

```
# R_t coefficients:
a02 <- function(t){-0.5*theta[1]*theta[1]}
c02 <- function(t){theta[1]*theta[1]}
d02 <- function(t){theta[1]*theta[4]*theta[5]}
# V_t coefficients:
b00 <- function(t){theta[2]}
b01 <- function(t){-theta[3]}
e02 <- function(t){theta[1]*theta[4]*theta[5]}
f02 <- function(t){theta[4]*theta[4]}

theta.start <- c(1,1,1,1,0)
model_3 <- BiGQD.mle(X,time,mesh=10,theta=theta.start)
```

```
## Compiling C++ code. Please wait.
```

```
## =====
##                      GENERALIZED QUADRATIC DIFFUSION
## =====
## ----- Drift Coefficients -----
## a02 : -0.5*theta[1]*theta[1]
## ... ..
## b00 : theta[2]
## b01 : -theta[3]
## ----- Diffusion Coefficients -----
## c02 : theta[1]*theta[1]
## ... ..
## d02 : theta[1]*theta[4]*theta[5]
## ... ..
## e02 : theta[1]*theta[4]*theta[5]
## ... ..
## f02 : theta[4]*theta[4]
##
## ----- Model Info -----
## Time Homogeneous      : Yes
## Data Resolution       : Homogeneous: dt=0.0192
## # Removed Transits.   : None
## Density approx.       : 4th Ord. Truncation, Bivariate-Saddlepoint
## Elapsed time          : 00:00:26
## ... ..
## dim(theta)            : 5
## -----
```

```
GQD.remove()
```

```
## [1] "Removed : a02 b00 b01 c02 d02 e02 f02"
```

```
# R_t coefficients:
a00 <- function(t){theta[1]}
a10 <- function(t){theta[7]}
a02 <- function(t){-0.5*theta[2]*theta[2]}
c02 <- function(t){theta[2]*theta[2]}
d02 <- function(t){theta[2]*theta[5]*theta[6]}
# V_t coefficients:
b00 <- function(t){theta[3]}
b01 <- function(t){-theta[4]}
e02 <- function(t){theta[2]*theta[5]*theta[6]}
f02 <- function(t){theta[5]*theta[5]}

theta.start <- c(0,1,1,1,1,0,0)
model_4 <- BiGQD.mle(X,time,mesh=10,theta=theta.start)
```

```
## Compiling C++ code. Please wait.
```

```
## =====
##                               GENERALIZED QUADRATIC DIFFUSION
## =====
## ----- Drift Coefficients -----
## a00 : theta[1]
## a10 : theta[7]
## a02 : -0.5*theta[2]*theta[2]
## ... ..
## b00 : theta[3]
## b01 : -theta[4]
## ----- Diffusion Coefficients -----
## c02 : theta[2]*theta[2]
## ... ..
## d02 : theta[2]*theta[5]*theta[6]
## ... ..
## e02 : theta[2]*theta[5]*theta[6]
## ... ..
## f02 : theta[5]*theta[5]
## -----
## ----- Model Info -----
## Time Homogeneous      : Yes
## Data Resolution       : Homogeneous: dt=0.0192
## # Removed Transits.   : None
## Density approx.       : 4th Ord. Truncation, Bivariate-Saddlepoint
## Elapsed time          : 00:00:21
## ... ..
## dim(theta)            : 7
## -----
```

```
GQD.remove()
```

```
## [1] "Removed : a00 a10 a02 b00 b01 c02 d02 e02 f02"
```

```
# R_t coefficients:
a00 <- function(t){theta[1]}
a02 <- function(t){-0.5*theta[2]*theta[2]}
c02 <- function(t){theta[2]*theta[2]}
d02 <- function(t){theta[2]*theta[5]*theta[6]}
# V_t coefficients:
b00 <- function(t){theta[3]}
b10 <- function(t){theta[7]}
b01 <- function(t){-theta[4]}
e02 <- function(t){theta[2]*theta[5]*theta[6]}
f02 <- function(t){theta[5]*theta[5]}
theta.start <- c(0,1,1,1,0,0)
model_5 <- BiGQD.mle(X,time,mesh=10,theta=theta.start)
```

```
## Compiling C++ code. Please wait.
```

```
## =====
##                      GENERALIZED QUADRATIC DIFFUSION
## =====
## ----- Drift Coefficients -----
## a00 : theta[1]
## a02 : -0.5*theta[2]*theta[2]
## ... ..
## b00 : theta[3]
## b10 : theta[7]
## b01 : -theta[4]
## ----- Diffusion Coefficients -----
## c02 : theta[2]*theta[2]
## ... ..
## d02 : theta[2]*theta[5]*theta[6]
## ... ..
## e02 : theta[2]*theta[5]*theta[6]
## ... ..
## f02 : theta[5]*theta[5]
## ----- Model Info -----
## Time Homogeneous      : Yes
## Data Resolution       : Homogeneous: dt=0.0192
## # Removed Transits.   : None
## Density approx.       : 4th Ord. Truncation, Bivariate-Saddlepoint
## Elapsed time          : 00:00:44
## ... ..
## dim(theta)            : 7
## -----
```

```
GQD.aic(list(model_1,model_2,model_3,model_4,model_5))
```

##	Convergence	p	min.likelihood	AIC	BIC	N
## Model 1	0	6	-7852.212	-15692.424	-15661.381	1305
## Model 2	0	6	-7965.957	[=] -15919.914	[=] -15888.871	1305
## Model 3	0	5	-7948.554	-15887.109	-15861.239	1305
## Model 4	0	7	-7966.154	-15918.308	-15882.090	1305
## Model 5	0	7	-7936.321	-15858.642	-15822.424	1305

`GQD.estimate(model_4)`

##	Estimate	Lower_95	Upper_95
## theta[1]	0.014	-0.394	0.422
## theta[2]	4.254	4.094	4.414
## theta[3]	0.094	0.066	0.122
## theta[4]	3.042	1.950	4.133
## theta[5]	1.746	1.683	1.809
## theta[6]	-0.647	-0.677	-0.616
## theta[7]	0.017	-0.043	0.077